## 3) Fruit Classification.

This question use the <u>fruit data</u> to determine which of the following fruits

- Banana
- Orange
- Other Fruit

are more likely to be Long, Not Sweet, and Not Yellow

Training Data							
Туре	_			Not Sweet	•	•	•
Banana Orange Other Fruit	400   0   100	100     300     100	350 150 150	150     150     50	450   300   50	50   0   150	500 300 200
Total	500		650	350   +	1	200	1000

## **Basic Probability**

• Prior probabilities

P (Banana) 
$$= 0.5 = 500/1000$$

P (Orange) 
$$= 0.3 = 300/1000$$

P (Other Fruit) = 
$$0.2 = 200/1000$$

• Probability of "Evidence"

$$P (Long) = 0.5 = 500/1000$$

P (Not Sweet) = 
$$0.35 = 350/1000$$

$$P \text{ (Not Yellow)} = 0.2 = 200/1000$$

Step 1 – Check Banana is more likely to be Long, Not Sweet, and Not Yellow.

By comparison, the formula of using **Bayes Theorem** to solve this problem is

P(Banana|Long, Not Sweet and Not Yellow)

=  $P(Banana|Long \cap Not Sweet \cap Not Yellow)$ 

P (Long 
$$\cap$$
 Not Sweet  $\cap$  Not Yellow | Banana) \* P(Banana)

= ------

P (Long 
$$\cap$$
 Not Sweet  $\cap$  Not Yellow)

```
= (count (Long, Not Sweet, Not Yellow, Banana)/count (Banana))
                     (count (Banana)/count(total))
        count(Long, Not Not Yellow, Banana) /count(total)
Note: It is hard to find count(Long, Not Not Yellow, Banana).
Applying Naive Bayes Formula -
P(Banana|Long, Not Sweet and Not Yellow)
   P(Long|Banana) * P(Not Sweet|Banana) * P(Not Yellow|Banana) * P(banana)
             _____
            P(Long) * P(Not Sweet) * P(Not Yellow)
  • P(Long|Banana) = count (Long, Banana)/count(Banana) = 400 / 500 = 0.8
  • P(Not Sweet|Banana)
      = count (Not Sweet, Banana) / count (Banana) = 150/500 = 0.3
  • P(Not Yellow|Banana)
      = count (Not Yellow, Banana) / count (Banana) = 50/500 = 0.1
Substitute Values in below Formula -
   P(Long|Banana) * P(Not Sweet|Banana) * P(Not Yellow|Banana) * P(banana)
=
            P(Long) * P(Not Sweet) * P(Not Yellow)
= (0.8*0.3*0.1*0.5)/ P(evidence) = 0.012/ P(evidence)
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## Step 2 – Check Orange is more likely to be Long, Not Sweet, and Not Yellow.

By comparison, the formula of using **Bayes Theorem** to solve this problem is P(Orange|Long, Not Sweet and Not Yellow)

```
= P(Orange | Long \cap Not Sweet \cap Not Yellow)
   P(Long \cap Not Sweet \cap Not Yellow | Orange) * P(Orange)
           P(Long \cap Not Sweet \cap Not Yellow)
= (count (Long, Not Sweet, Not Yellow, Orange)/count (Orange)) *
(count (Orange)/count(total))
      ______
        count(Long, Not Not Yellow, Orange) /count(total)
Note: It is hard to find count(Long, Not Not Yellow, Orange).
Applying Naive Bayes Formula -
P(Orange | Long, Not Sweet and Not Yellow)
  P(Long| Orange) * P(Not Sweet| Orange) * P(Not Yellow| Orange) * P(Orange)
              P(Long) * P(Not Sweet) * P(Not Yellow)
  • P(Long| Orange) = count (Long, Orange)/count(Orange) = 0/300 = 0
  • P(Not Sweet| Orange)
     = count (Not Sweet, Orange) / count (Orange) = 150/300 = 0.5
  • P(Not Yellow Orange)
     = count (Not Yellow, Orange) / count (Orange) = 0/300 = 0
Substitute Values in below Formula -
  P(Long| Orange) * P(Not Sweet| Orange) * P(Not Yellow| Orange) * P(Orange)
            _____
              P(Long) * P(Not Sweet) * P(Not Yellow)
```

```
= (0*0.5*0*0.3)/ P(evidence) = 0
```

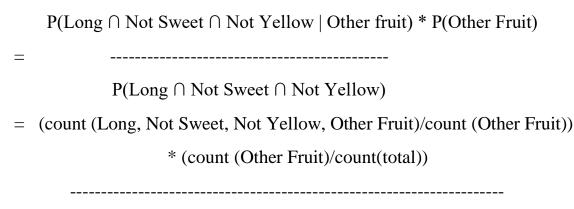
The value of P(Orange | Long, Not Sweet and Not Yellow) was zero in the above example, because, P(Long | Orange) and P(Not Yellow| Orange) was zero. That is, there were no 'Long' and 'Not Yellow' oranges in the training data.

It makes sense, but when you have a model with many features, the entire probability will become zero because one of the feature's value was zero.

Step 3 – Check Other Fruit is more likely to be Long, Not Sweet, and Not Yellow.

By comparison, the formula of using **Bayes Theorem** to solve this problem is P(Other Fruit |Long, Not Sweet and Not Yellow)

= P(Other Fruit | Long  $\cap$  Not Sweet  $\cap$  Not Yellow)



Note: It is hard to find count (Long, Not Sweet, Not Yellow, Other Fruit).

count(Long, Not Sweet , Not Yellow, Other Fruit) /count(total)

**Applying Naive Bayes Formula -**

P(Other Fruit | Long, Not Sweet and Not Yellow)

P(Long| Other Fruit) \* P(Not Sweet| Other Fruit) \* P(Not Yellow| Other Fruit) \* P(Other Fruit)

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P(Long) \* P(Not Sweet) \* P(Not Yellow)

• P(Long| Other Fruit) = count (Long, Other Fruit)/count (Other Fruit)

$$=100/200=0.5$$

- P(Not Sweet| Other Fruit)
  - = count (Not Sweet, Other Fruit) / count (Other Fruit) = 50/200 = 0.25
- P(Not Yellow Other Fruit)
  - = count (Not Yellow, Other Fruit) / count (Other Fruit) = 150/200 = 0.75

Substitute Values in below Formula -

P(Long| Other Fruit) \* P(Not Sweet| Other Fruit) \* P(Not Yellow| Other Fruit)

\* P(Other Fruit)

= ------

P(Long) \* P(Not Sweet) \* P(Not Yellow)

= (0.5\*0.25\*0.75\*0.2) / P(evidence) = 0.01875 / P(evidence)

The denominator is the same for all 3 cases, so it's optional to compute.

Clearly, Other Fruit gets the highest probability, so that will be our predicted class.

## Summary -

By an overwhelming margin (0.01875 >> 0.012), we classify this Not Sweet/Long/Not Yellow fruit as likely to be an Other Fruit.